

## EX1:

- a. Use EM algorithms and the generated samples to estimate the unknown parameters

$$m = [1 \ 1; 5 \ 5; 9 \ 1]'$$

$$S = [1 \ 0.4; 0.4 \ 1;$$

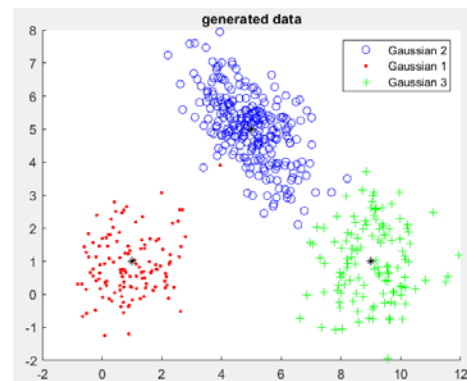
$$1 \ -0.6; -0.6 \ 1;$$

$$1 \ 0; 0 \ 1];$$

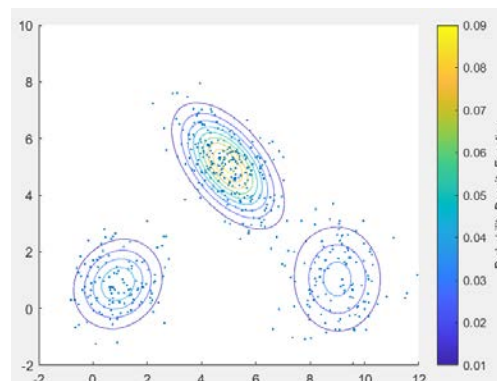
Generate 500 samples according “the first two samples are generated from the 2<sup>nd</sup> Gaussian, the 3<sup>rd</sup> sample from the 1<sup>st</sup> one, and the 4<sup>th</sup> sample from the last Gaussian”.

	1	2	3	4	5	6	7	8	9	10
1	6.1650	5.0751	0.3035	9.0591	5.2641	3.5538	2.2460	9.5774	4.8644	3.7296
2	4.8025	5.2362	2.2759	2.7971	5.5389	5.3068	0.9128	0.6400	4.0019	6.5499

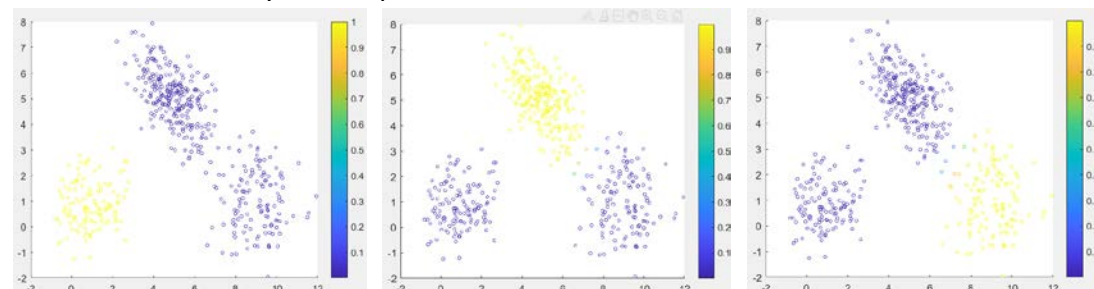
Generate ground truth data



Probability Density Function



Posterior Probability of Component 1, 2, 3



Result:

$$P1 = 0.25016$$

$$P2 = 0.49691$$

$$P3 = 0.25293$$

$$\mu_1 = \begin{bmatrix} 0.9428 & 0.9157 \end{bmatrix}$$

$$\mu_2 = \begin{bmatrix} 4.9957 & 5.0086 \end{bmatrix}$$

$$\mu_3 = \begin{bmatrix} 8.9792 & 0.8659 \end{bmatrix}$$

$$\sigma_1 = \begin{bmatrix} 0.9773 & 0.2663 \\ 0.2663 & 1.0275 \end{bmatrix}$$

$$\sigma_2 = \begin{bmatrix} 1.0868 & -0.5956 \\ -0.5956 & 0.9393 \end{bmatrix}$$

$$\sigma_3 = \begin{bmatrix} 0.7765 & 0.1923 \\ 0.1923 & 0.8832 \end{bmatrix}$$

## b. Same as above

$m = [1 \ 1; 3.5 \ 3.5; 6 \ 1]'$

$S = [1 \ 0.4; 0.4 \ 1;$

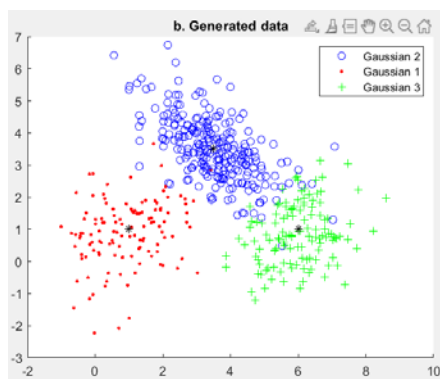
$1 \ -0.6; -0.6 \ 1;$

$1 \ 0; 0 \ 1];$

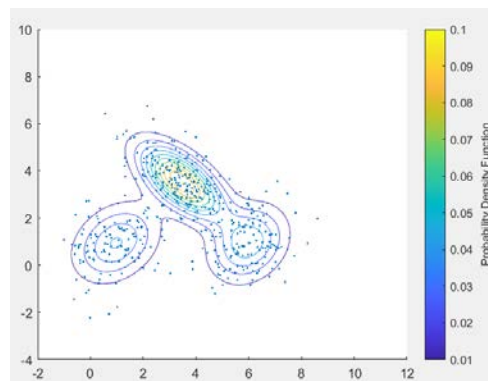
Generate 500 samples according “the first two samples are generated from the 2<sup>nd</sup> Gaussian, the 3<sup>rd</sup> sample from the 1<sup>st</sup> one, and the 4<sup>th</sup> sample from the last Gaussian”.

	1	2	3	4	5	6	7	8	9	10	11	12	13
1	4.0377	1.2412	1.3188	5.5664	7.0784	2.1501	1.7254	6.7147	3.3759	4.9090	1.6715	6.7172	3.9141
2	4.6445	5.5450	-0.0710	1.3426	3.5685	6.7379	1.2324	0.7950	4.7662	3.7883	0.1619	2.6302	4.0377

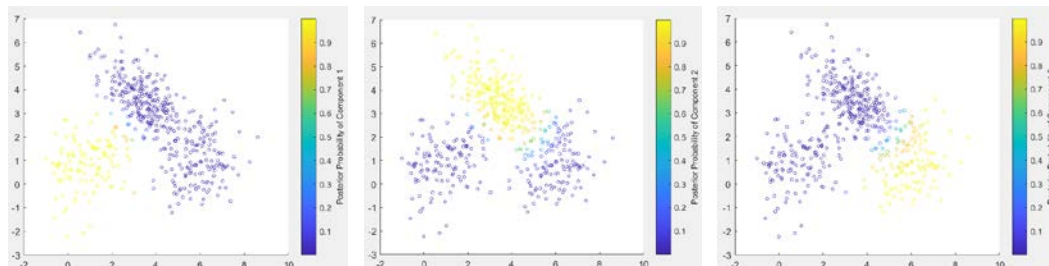
Generate ground truth data



Probability Density Function



Posterior Probability of Component 1, 2, 3



Result:

	$\mu_1 =$	$\sigma_1 =$
	0.9644    0.9475	1.0013    0.2997
$P_1 = 0.25446$	$\mu_2 =$	$\sigma_2 =$
$P_2 = 0.48436$	3.4751    3.5171	1.0113    -0.6312
$P_3 = 0.26117$	$\mu_3 =$	$\sigma_3 =$
	5.9760    0.9276	0.8011    0.1935
		0.1935    0.9825

c. Same as above

$m = [1 \ 1; 3.5 \ 3.5; 6 \ 1]'$

$S = [1 \ 0.4; 0.4 \ 1;$

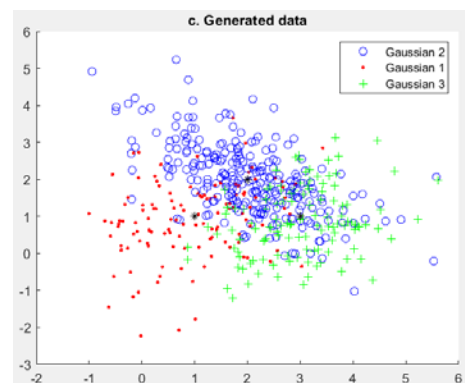
$1 \ -0.6; -0.6 \ 1;$

$1 \ 0; 0 \ 1];$

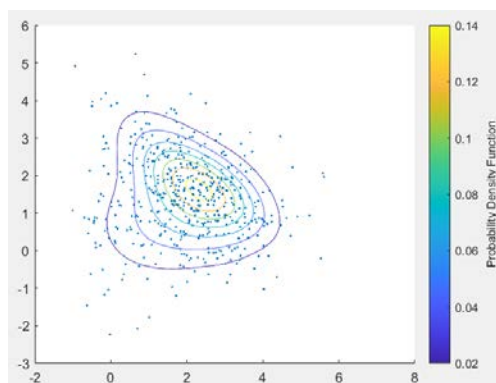
Generate 500 samples according “the first two samples are generated from the 2<sup>nd</sup> Gaussian, the 3<sup>rd</sup> sample from the 1<sup>st</sup> one, and the 4<sup>th</sup> sample from the last Gaussian”.

	1	2	3	4	5	6	7	8	9	10	11	12	13
1	2.5377	-0.2588	1.3188	2.5664	5.5784	0.6501	1.7254	3.7147	1.8759	3.4090	1.6715	3.7172	2.4141
2	3.1445	4.0450	-0.0710	1.3426	2.0685	5.2379	1.2324	0.7950	3.2662	2.2883	0.1619	2.6302	2.5377

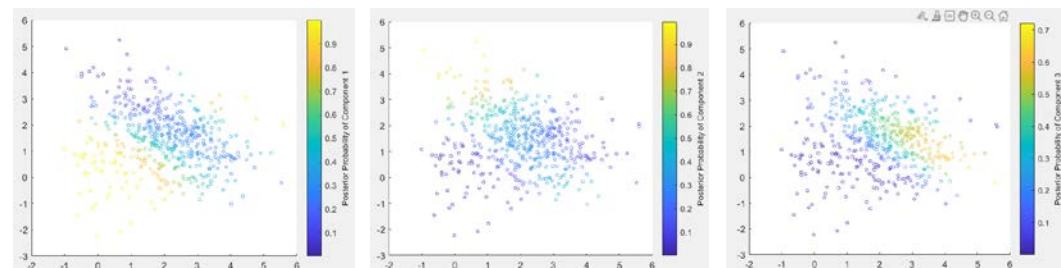
Generate ground truth data



Probability Density Function



Posterior Probability of Component 1, 2, 3



Result:

$P1 = 0.44565$	$\mu1 =$	$\sigma1 =$
	1.7474    0.9900	1.7980    0.2777
$P2 = 0.2924$	$\mu2 =$	0.2777    1.1010
	1.7912    1.9556	$\sigma2 =$
$P3 = 0.26195$	$\mu3 =$	1.1219    -0.8495
	2.6222    1.6183	-0.8495    1.5098
		$\sigma3 =$
		0.9813    -0.4760
		-0.4760    0.5891

In case of C. It's failed to converge in 100 iterations, which I have setting, so got the worse result.

d.

(2D-visualize as above.)

The original data set, which consists of three overlapping groups of points.

Trying to use Gaussian mixtures to get a more obvious clustering. Each one of the obtained clusters contains a significant percentage of points from more than one distribution.

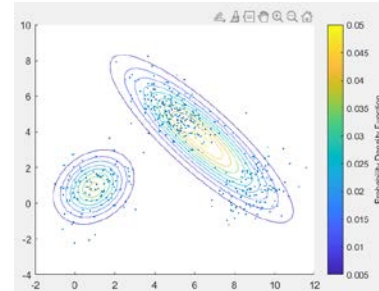
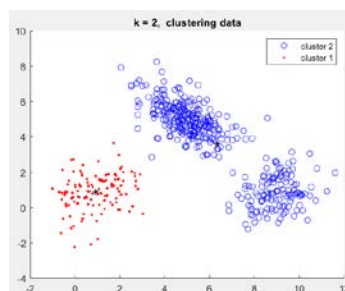
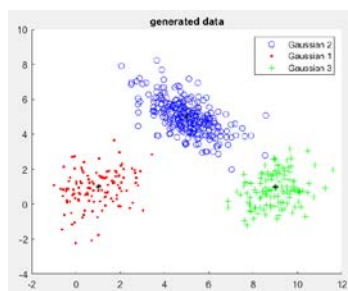
## EX2 K-means for

### (a) k=2

Generate ground truth data

k=2 clustering data

Probability Density Function



Result:

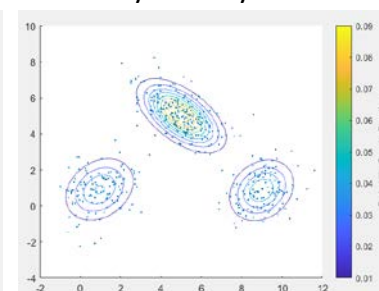
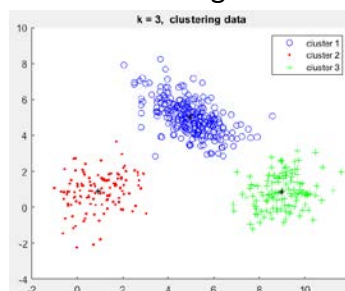
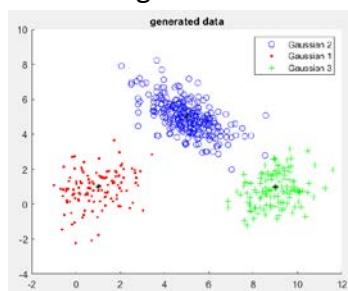
	mu1 =	sigma1 =
	0.9214	0.9686    0.2575
	0.8977	0.2575    1.0202
P1 = 0.24971	mu2 =	sigma2 =
P2 = 0.75029	6.3309	4.5320   -4.0155
	3.6094	-4.0155   4.7542

### (b) k=3

Generate ground truth data

k=3 clustering data

Probability Density Function



Result:

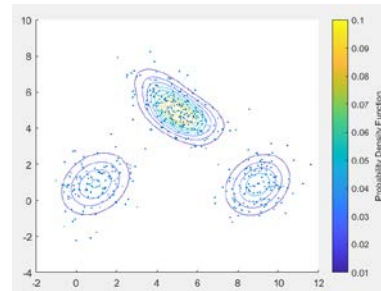
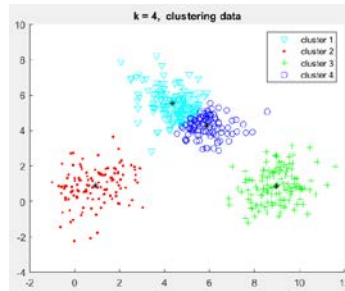
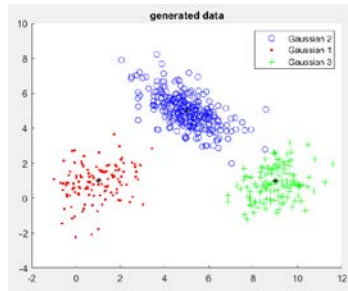
	mu1 =	sigma1 =
	0.9214	0.7765    0.1921
	0.8977	0.1921    0.8838
P1 = 0.25295	mu2 =	sigma2 =
P2 = 0.49689	4.9839	1.0865   -0.5954
P3 = 0.25016	5.0073	-0.5954   0.9392
	mu3 =	sigma3 =
	8.9720	0.9773   0.2663
	0.8688	0.2663   1.0275

### (c) k=4

Generate ground truth data

k=4 clustering data

Probability Density Function



Result:

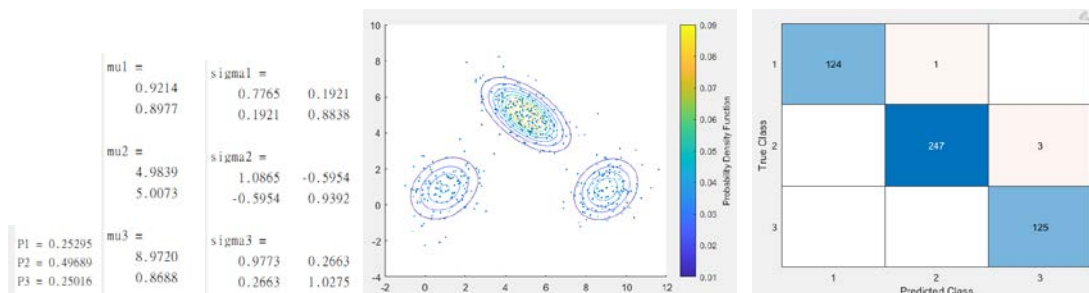
P1 = 0.29539 P2 = 0.25319 P3 = 0.2013 P4 = 0.25012	mu1 =	0.9214 0.8977	sigma1 =	0.9136 -0.6623 -0.6623 1.1224
	mu2 =	5.8878 4.2440	sigma2 =	0.7789 0.1909 0.1909 0.8754
	mu3 =	8.9720 0.8688	sigma3 =	1.1122 -0.2694 -0.2694 0.4512
	mu4 =	4.3670 5.5282	sigma4 =	0.9768 0.2651 0.2651 1.0255

### EX3 Fuzzy k-means algorithm (k=3)

#### (a) q=2

	1	2	3	4	5	6	7	8	9	10	11	12	13
1	0.0304	0.1613	0.9536	0.0052	0.0899	0.1447	0.9550	0.0078	0.0318	0.0402	0.9473	0.0374	0.0
2	0.9312	0.7514	0.0279	0.0114	0.5141	0.7465	0.0305	0.0149	0.9367	0.8815	0.0321	0.1052	0.9
3	0.0383	0.0872	0.0185	0.9834	0.3960	0.1088	0.0145	0.9773	0.0314	0.0783	0.0207	0.8574	0.0

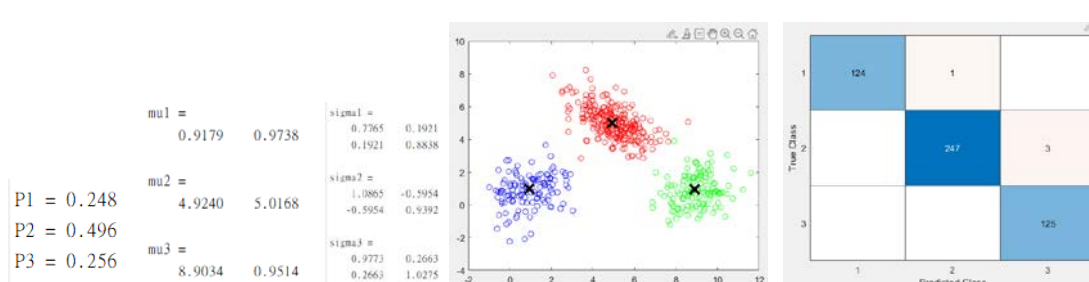
Confusion matrix:



#### (b) q=3

	1	2	3	4	5	6	7	8	9	10	11	12	13
1	0.1330	0.2591	0.7543	0.0578	0.1825	0.2436	0.7757	0.0768	0.1374	0.1424	0.7453	0.1338	0.0
2	0.7181	0.5513	0.1354	0.0855	0.4337	0.5459	0.1327	0.1058	0.7264	0.6588	0.1413	0.2238	0.7
3	0.1490	0.1896	0.1103	0.8567	0.3838	0.2105	0.0916	0.8175	0.1362	0.1988	0.1135	0.6423	0.1

Confusion matrix:



FCM 的群心是藉由 "所有的數據點" 以 "距離乘上隸屬值" 而成。其中  $q$  值稱為 fuzzier，就是 FCM 中的 Fuzzy，此值可以調控該項對於群心計算的影響力。